

**Numerical Methods and Computation**

**Prof. S.R.K. Iyengar**

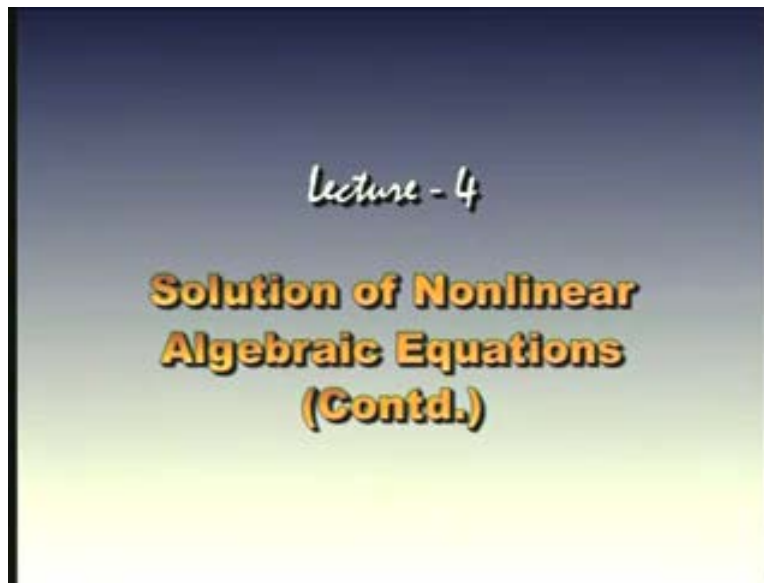
**Department of Mathematics**

**Indian Institute of Technology, Delhi**

**Lecture No # 4**

**Solution of Nonlinear Algebraic Equations (Continued)**

(Refer Slide Time: 00:00:56 min)



In our previous lecture we derived the Chebyshev method. The implementation of the method requires evaluation of three quantities. One is the function  $f(x)$ , the other is its derivative  $f'(x)$  and its second derivative  $f''(x)$ . Further we also mentioned that the method is of third order, which would mean that if you are away from the exact root by say 0.1 then each iteration can produce an accuracy of ten to the power of minus three such that the method is sufficiently fast for any requirement. However in many practical problems it may not be possible for us to find out the second derivative or even the first derivative, therefore in such circumstances we need an alternative method where in the evaluation of the derivatives may not be possible. So in that direction we have what is known as the Muller's method.

(Refer Slide Time: 00:03:05 min)

$$x_2 = x_1 - \frac{f_1}{f'_1} - \frac{1}{2} \left( \frac{f_1}{f'_1} \right)^2 \left( \frac{f''_1}{f'_1} \right) = 0.142854$$

$$\text{Exact value} = \frac{1}{7} = 0.142857$$

Müller method:

We assume for  $f(x)$  a polynomial of degree two in the form

$$f(x) = a_0(x - x_k)^2 + a_1(x - x_k) + a_2 = 0, a_0 \neq 0 \quad (9)$$

The Muller method also uses the concept of approximating the exact root by a quadratic equation. In the neighborhood of the root the curve is approximated by a quadratic equation. Now there are number of ways of writing a quadratic polynomial. One way that we used for deriving Chebyshev method was to write the polynomial  $a_0x$  square plus  $a_1x$  plus  $a_2$  is equal to zero. However an alternative way of writing a quadratic polynomial is  $a_0x$  minus  $x_k$  whole square plus  $a_1x$  minus  $x_k$  plus  $a_2$  is equal to zero, where we take  $a_0$  not equal to zero because it is a quadratic polynomial. Now if I open it up and simplify what I would get is some  $a_0x$  square plus some  $b_1$  into  $x$  plus  $b_2$  is equal to zero. Therefore this is a suitable form for me in order that the method can be derived much more easily.

(Refer Slide Time: 00:05:41 min)

Let  $x_{k-2}, x_{k-1}, x_k$  be any three approximations to the root.

Substituting  $x = x_k, x_{k-1}$  and  $x_{k-2}$ , we determine  $a_0, a_1$  and  $a_2$  from the equations

$$f_k = a_2$$

$$f_{k-1} = a_0(x_{k-1} - x_k)^2 + a_1(x_{k-1} - x_k) + a_2$$

$$f_{k-2} = a_0(x_{k-2} - x_k)^2 + a_1(x_{k-2} - x_k) + a_2$$

Now the application of the Muller's method requires three initial approximations. If you remember that the Newton Raphson required one initial approximation, Secant method required two initial approximations and the Chebyshev method required one initial approximation. But for implementing this Muller method we need three approximations to the exact root. These three roots are any three approximations near the root. The exact root need not lie between any one of these intervals. For example if the root is say 5, you could choose on one side of it; say 3, 3.5, 4, 4.5, 4.7 or arbitrarily one can choose the approximation. Once we know that the root lies between a certain intervals, any three points within this interval can be taken as the initial approximation to the root. Therefore I would take the three approximations to the root as  $x_k$ ,  $x_{k-1}$ ,  $x_{k-2}$ ; which means in order to start my implementation method I need  $x_0$ ,  $x_1$ ,  $x_2$  to start and then find  $x_3$ ,  $x_4$ ,  $x_5$  five from the method. Now since these are the three approximations, obviously they are three points on the curve i.e.  $x_k$   $f$  at  $x_k$ ,  $x_{k-1}$   $f$  at  $x_{k-1}$ ,  $x_{k-2}$   $f$  at  $x_{k-2}$ ; through three points we can always plot a quadratic curve, that is a parabola.

Now this parabola will intersect the axis and that point cuts the axis at two points, again out of which one root will be the exact and the other root will not be approximating the exact root. So what I would just substitute  $x_k$ ,  $x_{k-1}$ ,  $x_{k-2}$  in this which I had written in the previous step. So you see the reason why we had written will be obvious now. When I substitute  $x$  is equal to  $x_k$  in this, these two terms drop, so  $a_2$  two gets evaluated automatically. So  $a_2$  will be simply equal to  $f_k$ . Now substitute  $x_k$ ;  $x$  is equal to  $x_{k-1}$ , then I would get  $a_0 x_k$  minus one minus  $x_k$  whole square, then  $a_1$  into  $x_k$  minus one minus  $x_k$  plus  $a_2$ . Similarly I will substitute the approximation  $x_{k-2}$  here, so that I have  $f_k$  minus two. This is  $x_k$  minus two and this is  $x_k$  minus two minus  $x_k$  plus two is equal to zero. These are three equations and three unknowns;  $a_2$ ,  $a_0$  and  $a_1$ . Now since  $a_2$  is determined I can take  $a_2$  to the left hand side here, then I will have a system of two equations for determining  $a_0$  and  $a_1$ .

(Refer Slide Time: 00:06:00 min)

$$f_{k-1} = a_0(x_{k-1} - x_k)^2 + a_1(x_{k-1} - x_k) + a_2$$

$$f_{k-2} = a_0(x_{k-2} - x_k)^2 + a_1(x_{k-2} - x_k) + a_2$$

Solving these equations, we obtain

$$a_2 = f_k$$

$$a_1 = \frac{1}{D} [(x_k - x_{k-2})^2 (f_k - f_{k-1}) - (x_k - x_{k-1})^2 (f_k - f_{k-2})]$$

$$a_0 = \frac{1}{D} [(x_k - x_{k-2})(f_k - f_{k-1}) - (x_k - x_{k-1})(f_k - f_{k-2})]$$

(10)

where

$$D = (x_{k-1} - x_k)^2 (x_{k-2} - x_k) - (x_{k-2} - x_k)^2 (x_{k-1} - x_k)$$

So I would solve these equations. Now  $a_2$  is determined from this. So  $a_2$  is equal to  $fx_k$ . I have taken this  $a_2$  to the left hand side, then I am finding first of all  $D$  as the determinant of the coefficient matrix. So I evaluate the determinant coefficient matrix i.e.  $D$  that is  $x_k$  minus one. If I take the determinant of the coefficient, what I would have here is  $x_k$  minus one into  $x_k$  whole square into  $x_k$  minus one minus  $x_k$  minus the product of these two theorems i.e.  $x_k$  minus two minus  $x_k$  whole square into  $x_k$  minus one minus  $k$ . So this is the determinant of the coefficient matrix for  $a_0$ , determinant  $a_0 a_1$ . Therefore I can find out  $a_1$  by using the Cramer's rule. I have evaluated the numerator and similarly I evaluated for  $a_0$  - this is the numerator that comes from there. It was evaluated simply by using the Cramer's rule for two by two systems of equations. Therefore I am able to determine the constants  $a_0$  and  $a_1$  and  $a_2$  very easily by assuming the polynomial form in the special form that we have taken there.

If I had taken it as  $a_0 x^2$  plus  $a_1 x$  plus  $a_2$ , I will have to solve a system of three by three equations and the system of three by three equations would give me the same solution but in a different format wherein I will have to combine the number of terms to get this particular format. But we will see this format later on. We shall be using such an approximation for a polynomial degree to find out some other numerical methods also. Therefore once I determine  $a_0$ ,  $a_1$  and  $a_2$  term here, I would go back and substitute in this polynomial.

(Refer Slide Time: 00:07:57 min)

$$D = \begin{vmatrix} x_{k-1} - x_k & x_{k-2} - x_k \\ x_{k-2} - x_{k-1} & x_k - x_{k-1} \end{vmatrix} \quad (10)$$

where

$$D = (x_{k-1} - x_k)^2 (x_{k-2} - x_k) - (x_{k-2} - x_k)^2 (x_{k-1} - x_k)$$

$$D = (x_{k-1} - x_k)(x_{k-2} - x_k)(x_{k-1} - x_k - x_{k-2} + x_k)$$

$$= (x_k - x_{k-1})(x_k - x_{k-2})(x_{k-1} - x_{k-2})$$

Solving equation (9) for  $(x - x_k)$  and replacing  $x$  by  $x_{k+1}$ , we obtain

$$x_{k+1} = x_k + \frac{-a_1 \pm \sqrt{a_1^2 - 4a_0 a_2}}{2a_0}$$

$$= x_k - \frac{a_2}{a_1} \quad k = 2, 3, \dots$$

Before let us first of all simplify this  $D$  that we had written over here because this can be simplified further. So if I now simplify this, you can see that  $x_k$  minus one minus  $x_k$  is a common factor;  $x_k$  minus two minus  $x_k$  is a common factor. So I can take out both these common factors  $x_k$  minus one minus  $x_k$  and  $x_k$  minus two minus  $x_k$  and what is left out is the  $x_k$  minus one minus  $x_k$  and minus  $x_k$  minus two plus  $x_k$ . So  $x_k$  cancels. I have a factor of this particular form.

Now you can see that the denominator of this can easily be written down once we know the three approximations. You start with any approximation either  $x_k$  that means from  $x_k$  or the other end. So we have the three points,  $x_k$  minus two  $x_k$  minus one and  $x_k$ . The starting one is  $x_k$  minus two and the last one is the  $x_k$ . We can start with any one of them. So what I do is subtract  $x_k$  from the remaining two previous approximations which are  $x_k$  minus  $x_k$  minus one,  $x_k$  minus  $x_k$  minus two. Then I move to the next point in the order  $x_k$  minus one minus  $x_k$  minus two, so this will be the denominator. So you can look at it in the reverse way, you can start from the other end.

You start from  $x_k$  minus two. Let us take a minus sign throughout. So  $x_k$  minus two minus  $x_k$  minus one then  $x_k$  minus two minus  $x_k$ , that is over. So I have to move to the next point which is  $x_k$  minus one minus  $x_k$  with a negative sign. So if I move from that sign I will have a minus sign, so I can do from either side but it is always good that we start with the current approximation that is available like previous one that is  $x_k$ . So that we do not have to take care of any sign, so that is simply in the product in the denominator. Now since we have solved for  $a_1$ ,  $a_0$ ,  $a_2$  we can now go back and substitute in this. So I substitute it for  $a_0$ ,  $a_1$  and  $a_2$  here and once I substitute it over here I can write down the next approximation. So the next approximation is  $f$  of  $x_k$  plus one. By substituting this is equal to zero, I am putting  $x_k$  plus one here,  $x_k$  plus one here. This is a quadratic equation for  $x_k$  plus one minus  $x_k$ , so I solve by the ordinary method which is  $x_k$  minus one  $x_k$  plus one minus  $x_k$  is equal to minus one plus minus under root  $a_1$  square four times  $a_0 a_2$  divided by two times  $a_1$ . So I have just solved this equation for  $x_k$  plus one minus  $x_k$  that is I have taken  $x_k$  to the right hand side. So it is trivial, so finding the root of this quadratic equation and both the roots we are taking over here. This particular form can be simplified by rationalizing this.

(Refer Slide Time: 00:10:42 min)

$$\begin{aligned}
 D &= (x_{k-1} - x_k)(x_{k-2} - x_k)(x_{k-1} - x_k - x_{k-2} + x_k) \\
 &= (x_k - x_{k-1})(x_k - x_{k-2})(x_{k-1} - x_{k-2})
 \end{aligned}$$

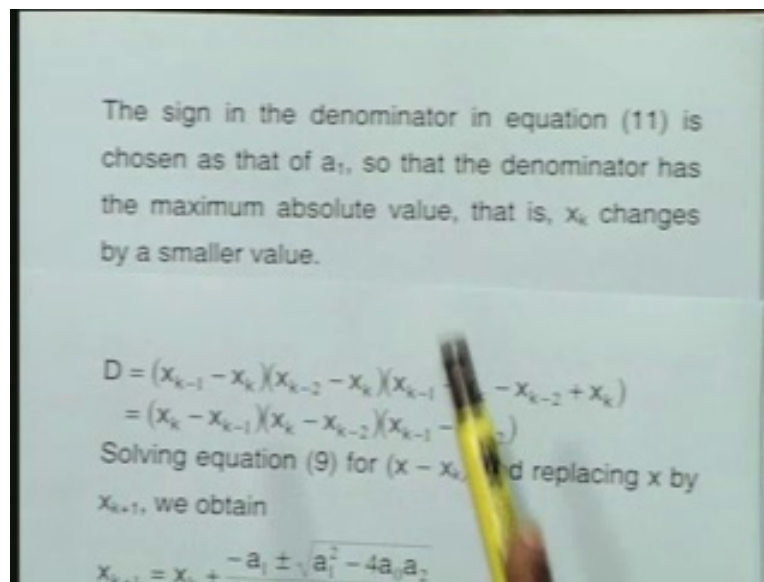
Solving equation (9) for  $(x - x_k)$  and replacing  $x$  by  $x_{k+1}$ , we obtain

$$\begin{aligned}
 x_{k+1} &= x_k + \frac{-a_1 \pm \sqrt{a_1^2 - 4a_0a_2}}{2a_0} \\
 &= x_k - \frac{2a_2}{a_1 \pm \sqrt{a_1^2 - 4a_0a_2}}, \quad k = 2, 3, \dots \quad (11)
 \end{aligned}$$

I multiply the numerator and denominator by  $a_1$  plus minus under root of  $a_1$  square minus four  $a_0$ ,  $a_0$ . When I multiplied it you could see  $a_1$  square cancels with  $a_1$  square and I have a plus minus four  $a_0$   $a_2$ ; out of this four  $a_0$  there is two  $a_0$  in the denominator, it cancels. So what is left out is simply two times  $a_2$  two and in the denominator I have  $a_1$  plus minus under root of  $a_1$  square minus four  $a_0$   $a_2$ . So the minus sign has been taken care of by writing this. Now there is a reason for writing this rationalization in this particular form.

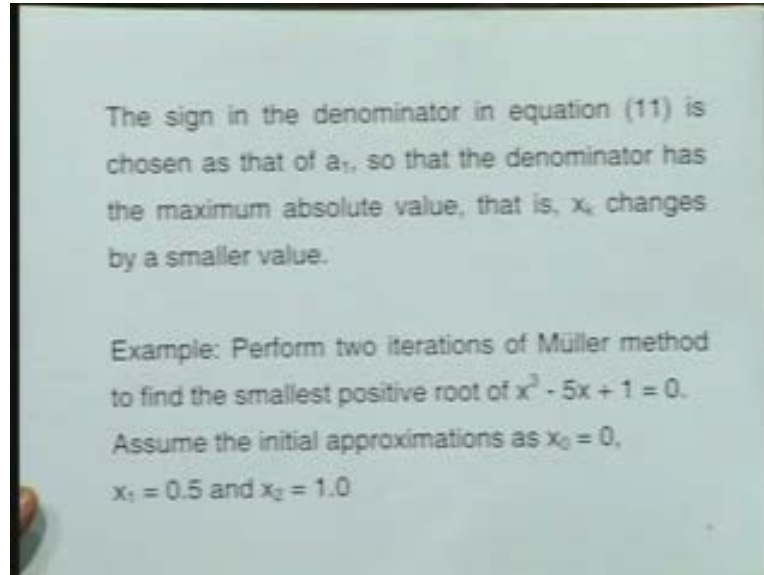
We have started with initial approximations  $x_k$  minus two,  $x_k$  minus one and  $x_k$ . Since the solution is a continuous solution the next root that we are going to get or next approximation we are going to get should be closer to  $x_k$ . It cannot jump far away from the  $x_k$ . Therefore what we do here is in order to choose the correct value to be added to  $x_k$ , I would see that the change that comes here is small; because there are two values here. I will choose that particular value which gives me the smaller of two in magnitude.

(Refer Slide Time: 00:11:58 min)



So what I would do is the sign in the denominator in equation eleven is chosen as that of  $a_1$ , so that the denominator has a maximum absolute value, that is,  $x_k$  changes by smaller value. Let me explain it again. We are taking the square root; therefore this is a positive square root. Now this quantity will be small if the denominator is large in magnitude. Therefore depending on the sign of  $a_1$ , if  $a_1$  is positive I will take positive sign here, then the denominator is the largest value. If  $a_1$  is negative, I will choose the negative value here, so that in magnitude it is large and in this quantity it will be smaller in magnitude.

(Refer Slide Time: 00:12:39 min)

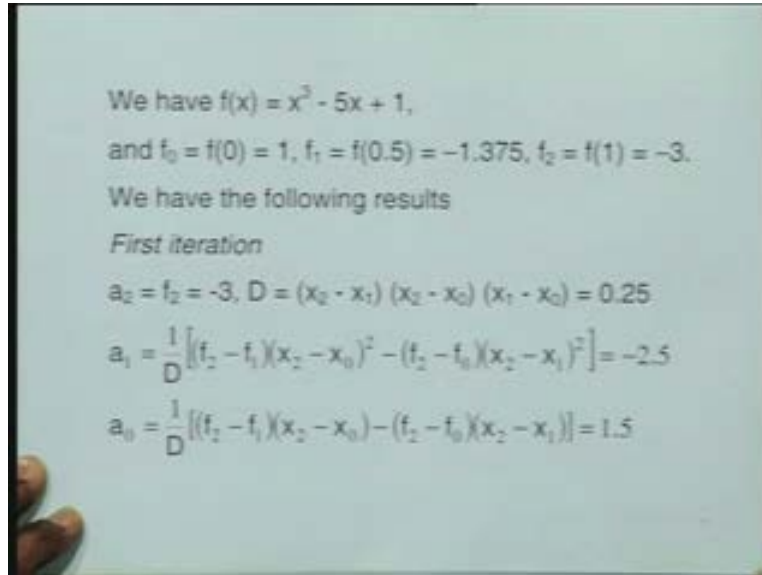


We discussed that the sign in the denominator in this equation is chosen as that of  $a_1$ , so that the denominator has the maximum absolute value, that is,  $x_k$  changes by the smallest possible value. This will automatically determine throwing away of the other root because the fact that we had rationalized it and brought in this particular form, we are automatically throwing away the second root by saying the other root is going to give a bigger change to  $x_k$ . Therefore that is how we have solved the problem.

Now let us discuss what we have to do for the next iteration, what is the computational cost per iteration. We have the approximations  $x_{k-1}$ ,  $x_{k-2}$ ,  $x_{k-1}$  and  $x_k$ . We have computed  $x_{k+1}$ . Now we go to the next step where I need evaluation of  $f(x_{k+1})$ . I do not need anything else here because  $a_0$ ,  $a_1$  and  $a_2$  do not contain any other thing. Therefore the cost of evaluation is simply one function evaluation i.e.  $f$  of  $x_{k+1}$ , so the cost of evaluation is only one function vanishing. However we shall show later on that the method is not quadratic or cubic. It drops down a super linear at about 1.8 which is superior to the secant method but it will only be 1.8, as the rate of convergence. We state here that what we have derived here is not superior to Chebyshev method or superior to the Newton Raphson method.

Now let us take this example. Perform two iterations of the Muller method to find the smallest positive root of  $x^3 - 5x + 1 = 0$  and here in the problem the initial approximations are given to us as 0.5 and 1. We already know and we have shown that the roots lies between 0 and 1 so we have just taken three points randomly 0, 0.5 and 1. If it so happens, depending on the root if you are able to take a different approximation, you may reduce one or two iterations and if it is not, it will take one or two more iterations to adjust and converge from one side.

(Refer Slide Time: 00:15:00 min)



We have  $f(x) = x^3 - 5x + 1$ ,  
and  $f_0 = f(0) = 1$ ,  $f_1 = f(0.5) = -1.375$ ,  $f_2 = f(1) = -3$ .  
We have the following results  
*First iteration*  
 $a_2 = f_2 = -3$ ,  $D = (x_2 - x_1)(x_2 - x_0)(x_1 - x_0) = 0.25$   
 $a_1 = \frac{1}{D} [(f_2 - f_1)(x_2 - x_0)^2 - (f_2 - f_0)(x_2 - x_1)^2] = -2.5$   
 $a_0 = \frac{1}{D} [(f_2 - f_1)(x_2 - x_0) - (f_2 - f_0)(x_2 - x_1)] = 1.5$

Now let us solve this problem. Now here we have  $f(x)$  is equal to  $x^3 - 5x + 1$ . Now I evaluate  $f$ . I have taken the three approximations, which are  $f$  at zero i.e.  $f_0$ ,  $f$  at point five i.e.  $f_1$ ,  $f$  at one is  $f_2$ . So these three values are 1, minus 1.375 and minus 3. Now we will substitute it and then get first iteration. Now if you remember that  $a_2$  is computed as  $f_2$  i.e. minus 3, then we said the denominator is starting with  $x_2$ ,  $x_2$  minus  $x_1$ ,  $x_2$  minus  $x_0$ ,  $x_1$  minus  $x_0$ . So I am starting from  $x_2$  and then taking all the subtractions from all the remaining approximations and this I can compute easily and it comes out as 0.25. Once  $D$  the denominator is obtained then the numerator also forms really a particular way of writing it. You can see the way in which the sequence goes in the numerator also. This is  $f_2$  minus  $f_1$ ,  $x_2$  minus  $x_0$  whole square,  $f_2$  minus  $f_0$ ,  $x_2$  minus  $x_1$  whole square. So this  $f_2$ ,  $x_2$ ,  $f_2$ ,  $x_2$ ,  $f_1$  opposite value,  $f_0$  opposite value, this is how a particular sequence can also be remembered in writing even  $a_1$  one and  $a_2$ . So I can substitute this  $f_2$  and  $f_1$  here;  $x_2$  and  $x_0$  here;  $f_2$   $f_0$  here;  $x_2$   $x_1$  here, on simplifying this, I get minus 2.5. Similarly this  $a_0$  is  $f_2$  minus  $f_0$   $x_2$  minus  $x_0$  and there is no squares except that both these quantities are same except there are no squares here and this quantity is 1.5. So I have now computed  $a_0$ ,  $a_1$  and  $a_2$ . Once I compute  $a_0$ ,  $a_1$  and  $a_2$  I would then try to evaluate the quantity.

(Refer Slide Time: 00:16:55 min)

Since  $a_1 < 0$ , we get

$$x_3 = x_2 - \frac{2a_2}{a_1 - \sqrt{a_1^2 - 4a_0a_2}} = 1 - \frac{6}{2.5 + \sqrt{24.25}}$$

$$= 0.191857$$

Second iteration

$$a_2 = f_3 = 0.047777,$$

$$D = (x_3 - x_1) (x_3 - x_2) (x_2 - x_1) = 0.124512$$

Now  $a_1$  is less than zero. So you can see that  $a_1$  is minus 2.5. Therefore  $a_1$  is negative. Therefore in the denominator I will take the negative sign here out of this plus minus. So since  $a_1$  is negative, I take  $x_3$  as  $x_2$  minus two times  $a_2$  two; this is  $a_1$  with a negative sign minus under root  $a_1$  square minus four  $a_0a_1a_2$  and this is a simple computation. I can substitute the values of  $a_0$ ,  $a_1$  and  $a_2$  and get 0.191857. Now this gives us the first step of Muller iteration i.e. first iteration of the Muller method. Now I repeat the same thing and I go find out what is my  $a_2$ ;  $a_2$  is now  $f$  of  $x_3$ . The current estimate is  $x_3$ ; so  $f$  of three. I compute the value of  $f_3$  and then  $D$ . Now look at this  $x_3$ . I could have written this first  $-x_3$  minus  $x_2$ ,  $x_3$  minus  $x_1$  into  $x_2$  minus  $x_1$ . So now we need three approximations, we have not come to  $x_3$ , so we drop the original  $x_0$ . So we have  $x_1$ ,  $x_2$  and  $x_3$ . So I take  $x_3$  minus  $x_2$ ,  $x_3$  minus  $x_1$  i.e.  $x_3$  minus  $x_2$ ,  $x_3$  minus  $x_1$ . Now I move further and get  $x_2$  minus  $x_1$ , so that will be the denominator and this product is simply equal to this.

Now again I compute my  $a_1$ ;  $a_1$  is one upon  $D$ ,  $f_3$  minus  $f_2$ . Now you see here  $x_3$  minus  $x_1$  whole square,  $f_3$  minus  $f_1$ ,  $x_3$  minus  $x_2$  whole square and  $a_0$  is  $f_3$  minus  $f_2$   $x_3$  minus  $x_1$   $f_3$  minus  $f_1$   $x_3$  minus  $x_2$  and I get this. Now these are the two computed values for  $a_1$  and  $a_0$  which we have written from the original expression that we had here earlier. Again I find  $a_1$  is negative. Therefore I will have to take in denominator the negative sign, so that the change in  $x_k$ , the current value  $x_3$  is smaller.

(Refer Slide Time: 00:18:31 min)

$$\begin{aligned}a_1 &= \frac{1}{D} [(f_3 - f_2)(x_3 - x_1)^2 - (f_3 - f_1)(x_3 - x_2)^2] \\&= -5.138588 \\a_0 &= \frac{1}{D} [(f_3 - f_2)(x_3 - x_1) - (f_3 - f_1)(x_3 - x_2)] \\&= 1.691854 \\&\text{Since } a_1 < 0, \text{ we get}\end{aligned}$$

So I would therefore write  $x_4$  as  $x_3$  minus twice  $a_2$ ,  $a_1$  minus under root of  $a_1$  square minus four,  $a_0 a_2$ . Now I have just written the values of  $x_3$ ,  $a_2$  here,  $a_1$  with a negative sign which has been taken care of as a plus sign here and this has become plus under root of this and evaluation of this reduces to this value. Now this completes the evaluation of two iterations, we have done  $x_3$  and this is the value that we have got as  $x_4$ .

(Refer Slide Time: 00:19:34 min)

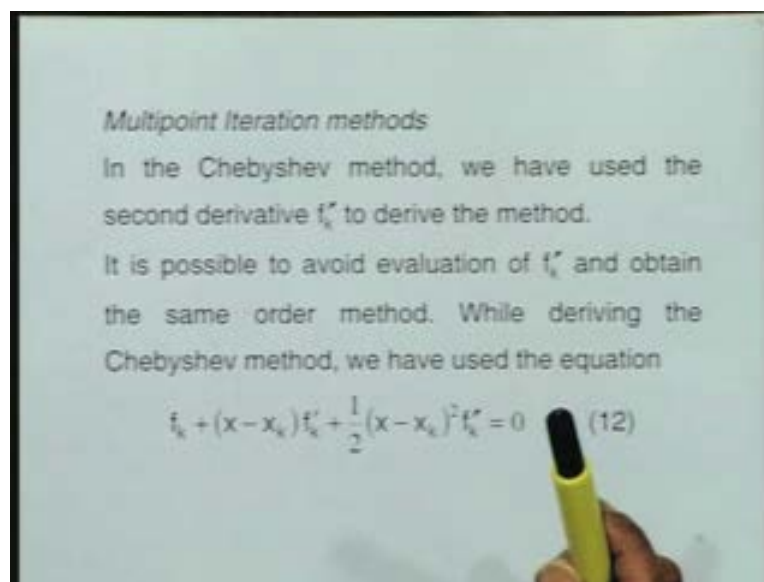
$$\begin{aligned}x_4 &= x_3 - \frac{2a_2}{a_1 - \sqrt{a_1^2 - 4a_0a_2}} \\&= 0.191857 + \frac{0.095554}{5.138588 + \sqrt{26.081760}} \\&= 0.201183\end{aligned}$$

Now as I mentioned earlier the reason why we have derived Muller's method is because, in practical applications it may not be possible for us to find out the second derivative as well as the

first derivative. Therefore we opted for a method which may not be as fast as the Chebyshev method but reasonably fast whose order of accuracy is 1.8 or rate of convergence is 1.8.

Now it may be possible that we must give an alternative that it may be possible for us to find the first derivative but not the second derivative. For that we have already Newton Raphson which uses first derivative. But if it is possible for me to get first derivative at some other point also it may be possible for me to get the same order as the Chebyshev method which is a third order method, that means it will work much superior to this. Therefore what we would like to do now is to start with the Chebyshev step but the way we have manipulated Chebyshev method; the Chebyshev method was obtained starting with the Taylor series expansion up to the second order, we drop the third order terms and then from there we manipulate it to get the Chebyshev method. However we would like to show now that if I manipulate it in a different way, I can avoid the evaluation of second derivative and I will evaluate one more first derivative which means I will have two first derivative evaluations and one function evaluation, at the same time achieve the third order accuracy which is order of the Chebyshev method.

(Refer Slide Time: 00:21:38 min)



Such methods are called multipoint iteration methods. They are simple manipulations. Why it is called multipoint is trivial; we are evaluating  $f'$  at two points, so it is a multipoint. More than one point it is being evaluated. So in a Chebyshev method we have used the second derivative. Now to avoid the evaluation of  $f''$  and obtain the same order method we shall use the following method. So we will start from where we have started our Chebyshev method. In the Chebyshev method we have written the Taylor expansion. We took the first three terms, the order of  $h^3$  term or order of  $(x - x_k)^3$  term is dropped. So from this we have manipulated the Chebyshev method. We are going to give multipoint iteration methods and both of them start from the same step.

(Refer Slide Time: 00:22:35 min)

Write it as

$$(x - x_k) \left[ f'_k + \frac{1}{2}(x - x_k)f''_k \right] = -f_k$$

Now, write an approximation as

$$x_{k+1} - x_k = - \frac{f_k}{f'_k + \frac{1}{2}(x_{k+1} - x_k)f''_k}$$

$$= - \frac{f_k}{f' \left[ x_k + \frac{1}{2}(x_{k+1} - x_k) \right]}$$

Now we write this step like this. What I do is I take  $f_k$  to the right hand side, retain these two terms on the left hand side, take  $x - x_k$  as common factor. So what is left out is  $x - x_k f'_k + \frac{1}{2}(x - x_k)^2 f''_k = -f_k$ . So it is simply taking this right hand side and taking the common factor. Then I would write from here the next approximation. I will take this factor to the denominator here and write down my next approximation. Therefore I will write  $x_k + 1 - x_k = f_k$  divided by this entire factor  $f'_k + \frac{1}{2}(x_{k+1} - x_k)f''_k$ . Now what I would do is the manipulation which we talked of will be done here. I will write this as the first two terms of the Taylor expansion of this. Let's look at this quantity here,  $f'_k + \frac{1}{2}(x_{k+1} - x_k)f''_k$ . The Taylor expansion of this is,  $f'_k$  at  $x_k$  plus this factor into next derivative i.e. second derivative. So this is half  $x_k$  plus one  $x_k$ . Let us even think this as  $h$ ; if you think this as  $h$ , this is simply  $h$  into  $f''_k$ . So that means these two terms get absorbed in the first two terms of the Taylor expansion of  $f'_k$  at  $x_k$ . Therefore I have now avoided taking the evaluation of  $f''_k$  but I have now brought it in this form, so that  $f''_k$  is being evaluated at one more value besides  $f''_k$  at  $k$ . Therefore the justification of calling a multipoint iteration that,  $f'_k$  is now being evaluated at a new value one more value besides the value at  $f'_k$ .

Now as we have done in the Chebyshev method we had  $x_k + 1$  on the right hand side,  $x_k + 1 - x_k$  was approximated by the Newton Raphson method and the method retained the order of three. We shall do the same thing here,  $x_k + 1 - x_k$  and we shall approximate it by the Newton Raphson method i.e.  $-f_k / f'_k$ .

(Refer Slide Time: 00:2:15 min)

$$x_{k+1} - x_k = - \frac{f_k}{f'_k + \frac{1}{2}(x_{k+1} - x_k)f''_k}$$

$$= - \frac{f_k}{f\left[x_k + \frac{1}{2}(x_{k+1} - x_k)\right]}$$


---


$$x_{k+1} = x_k - \frac{f_k}{f\left[x_k - \frac{1}{2}(f_k/f'_k)\right]}$$

For computation purposes, we write it as the two step method

So we will now approximate this quantity  $x_{k+1} - x_k$  by  $f_k$  minus  $f'$  prime  $k$ . So we simply are approximating this by Newton Raphson method so this is being replaced by minus  $f_k$  minus  $f'$  dash  $k$ . This method is of the same order as the Chebyshev method. The order does not fall down. The order is three. Now this is a multipoint iteration method, but this can be written in a nice fashion what we call as two step method. That means we shall write this method in two steps. One is evaluation of this particular argument of  $f'$  dash  $k$  and then evaluation of this. So we would therefore write this method as follows.

(Refer Slide Time: 00:26:39 min)

$$x_{k+1}^* = x_k - \frac{f_k}{2f'_k}$$

$$x_{k+1} = x_k - \frac{f_k}{f'_{k+1}}, \quad f'_{k+1} = f'(x_{k+1}^*) \quad (13)$$

This gives a multipoint iteration method in which  $f(x)$  is evaluated at two points.

We can also manipulate (12) as

So as you can see, I am now denoting this quantity that we have here as  $x_k$  plus one star as some intermediate value,  $x_k$  minus half  $f_k$  of dash  $k$  and then this method will be written as  $x_k$  plus one is  $x_k$  minus  $f_k$ , evaluation of  $f$  prime at this, so that this  $f$  prime is being evaluated as  $f$  prime at  $x$  star  $k$  plus one this one. So this almost looks like Newton-Raphson method except the other half here. If there was no half here it would have been a Newton-Raphson method. So this particular quantity is being used here and we compute  $x_k$  plus one star, then compute  $f$  prime at this particular argument and then we substitute it here and get the multipoint iteration method. Therefore this is called the multipoint iteration method, as one way of writing it.

Now the starting point for manipulation of this was this step. So we have manipulated this and brought this one step. If I manipulate it in a slightly different way I can get a different multipoint iteration method. Now let us see how we manipulate this particular thing here.

(Refer Slide Time: 00:27:43 min)

$$(x - x_k)f_k' = -\left[f_k + \frac{1}{2}(x - x_k)^2 f_k''\right]$$

$$= -[f(x_k + (x - x_k))] - (x - x_k)f_k'$$

The next approximation is written as

$$x_{k+1} - x_k = -\frac{1}{f_k'} [f(x_k + (x_{k+1} - x_k))] - (x_{k+1} - x_k)f_k'$$

$$= -\frac{1}{f_k'} \left[ f\left(x_k - \frac{f_k}{f_k'}\right) + \left(\frac{f_k}{f_k'}\right) f_k' \right]$$

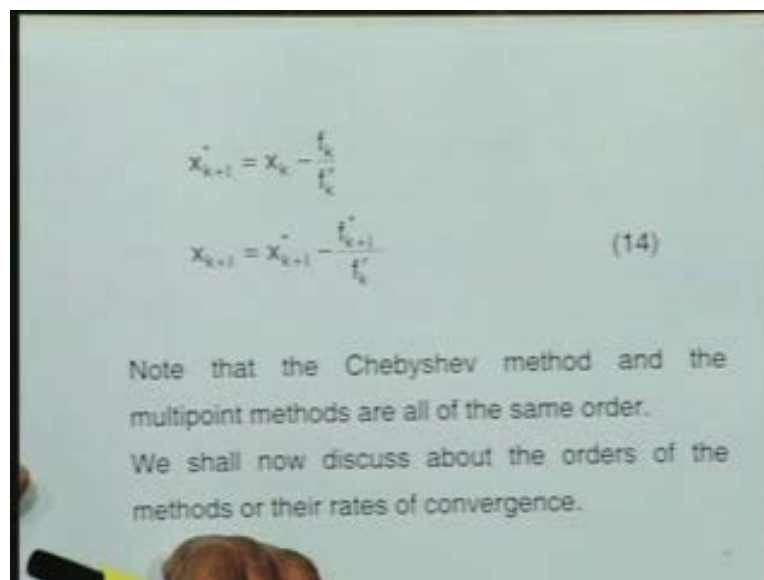
Now I retain the middle factor here and take these two to the right hand side, so  $x_k$  plus one minus  $x_k$   $f$  prime  $k$  is retained on the left hand side. I take  $f_k$  half  $f_k$  minus one to the whole square to the right hand side. Then the manipulation that I want to do is on the right hand side. So I would now try to manipulate this particular thing to avoid my second derivative. Let us forget for this moment, let us take this  $f$  of  $x_k$  plus  $x$  minus  $x_k$ ;  $x_k$  and  $x_k$  cancel each other. So if I open it, it gives you  $f$  at  $x_k$  plus  $x$  minus  $x_k$   $f$  prime of  $x_k$  which I have subtracted. So  $x$  minus  $x_k$   $f$  dash  $k$  cancels with  $x$  minus  $x_k$   $f$  prime  $x_k$  and the next term is half  $x$  minus  $x_k$  whole square  $f$  double dash  $k$  and that is this term. So to get this what we have done is we have added and subtracted  $x$  minus  $x_k$   $f$  dash  $k$  and these three terms have become the first three terms of the Taylor expansion of  $f$   $x_k$  plus  $x$  minus  $x_k$ .

Now if you now look at this one the this is now being evaluated at a new point,  $x_k$  plus  $x$  minus  $x_k$ . Now once I am able to write this, I can immediately write down the next approximation replacing  $x$  by  $x_k$  plus one. So this is  $x_k$  plus one minus  $x_k$ . I bring  $f$  prime  $k$  to the right hand side, so minus of one upon  $f$  prime  $k$   $f$   $x_k$  plus  $x_k$  minus one minus  $x_k$  and this  $x$  also is replaced

by  $x_k$  plus one. As we have done in the Chebyshev method as well as the previous multipoint iteration method, I would replace this  $x_k$  plus one minus  $x_k$  by the Newton Raphson method again. So this will this part will be read as  $f_k$  minus divided by  $f'$  prime  $k$  and this will read as  $f_k$  by  $f'$  prime  $k$ .

So let us simplify this one step further. Now if I simplify this I take a  $x_k$  to the right hand side then you look at the second term first;  $f'$  prime  $k$  cancels with the  $f'$  prime  $k$ , there is  $f_k$  here  $f'$  prime  $k$  here. So the second term is  $f_k$  minus  $f'$  prime  $k$  and third term is minus of  $f_k$  minus  $f'$  prime  $k$  divided by  $f'$  prime  $k$ . So this is another way of manipulating the starting approximation of the Chebyshev method to arrive at this particular new method. Now here again we have used the Newton Raphson method. Now I would like to write this also as a two step method.

(Refer Slide Time: 00:32:04 min)



The slide contains the following content:

$$x_{k+1}^* = x_k - \frac{f_k}{f'_k}$$

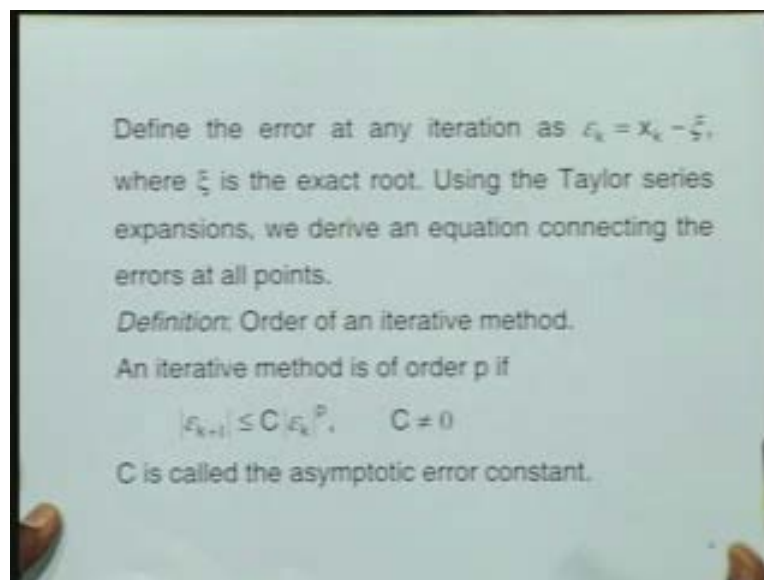
$$x_{k+1} = x_{k+1}^* - \frac{f_{k+1}}{f'_k} \quad (14)$$

Note that the Chebyshev method and the multipoint methods are all of the same order.  
We shall now discuss about the orders of the methods or their rates of convergence.

Now what I would do here is I would first of all call this a new approximation. This is  $x$  star  $k$  plus one is  $x_k$  minus  $f_k$  by  $f'$  dash  $k$ . Then I would write this  $x_k$  plus one is  $x$  star  $k$  plus one minus  $f$  star  $k$  plus one divided by  $f'$  dash  $k$ . A very interesting thing here is the first step is nothing but Newton Raphson method. So this is indeed a modification of the Newton Raphson method to arrive at the next higher order method. We know this will be a second order value and now I am computing a third order value. Here you can see that we are using two evaluations of  $f$ . We are using  $f$  star  $k$  plus one. Therefore I am now using here the evaluation of  $f$  two times; whereas in the previous multipoint iteration we had two evaluation of  $f'$  prime. So even though both of them are having the same number of evaluations (three evaluations) but the choice is with us whether the evaluation of  $f$  is easy for me or evaluation of  $f'$  prime is easy for me. If the evaluation of  $f$  is easy for me, I would chose this because what I am using is  $f$  star at  $k$  plus one. If evaluation of  $f'$  prime is easy I will use the other multipoint iteration method. Therefore this is exactly a modification Newton Raphson method to get a new method.

Now we have all along been saying that the order of Newton Raphson method is two order; order of Secant method is 1.6; order of Chebyshev method is 3. How do you get about it? The orders or rate of convergence of all these methods is very simple and very straight forward. Every thing depends on Taylor series. We define the error at any particular step, that means the current numerical solution minus exact solution is the error. Based on that I will substitute the expressions of that in any formula that I have got. I would bring any quantity that is there in the denominator. I will open up by Taylor series first then, whatever quantity is there in the denominator I will take it up and put it to the power of minus one, open it up as a binomial series, multiply everything and then cancel off. Whatever is left out will be an expression connecting the error i.e. the current step which is  $k$  plus one with the error at  $k$ ,  $k$  minus one and  $k$  minus two depending on the method that we are using.

(Refer Slide Time: 00:33:30 min)



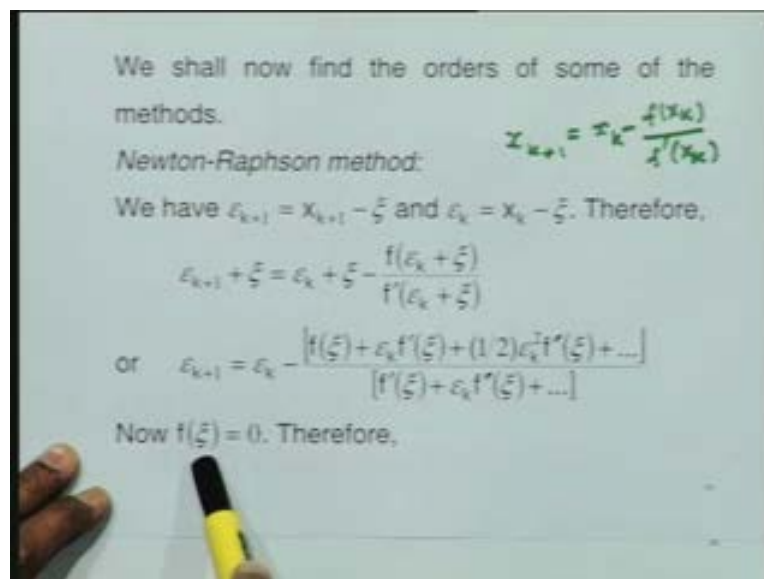
So to define this, let us first define the error as -  $x_k$  is a current approximation. the approximation minus  $\xi$  as  $\epsilon_k$ . Similarly  $\epsilon_{k+1}$  will be defined as  $x_{k+1}$  minus  $\xi$ ,  $\epsilon_{k+1}$  is equal to  $x_{k+1}$  minus  $\xi$  and so it depends upon the values that  $x_k$  are using. If you are using Newton Raphson we have only one value  $x_k$ ; if you are using Secant method we have to use  $x_k$  and  $x_{k-1}$ ; if you are using Muller method we have to use  $x_k$ ,  $x_{k-1}$  and  $x_{k-2}$ .

As I mentioned how the order of the method is defined by connecting how the error at the present step is connected to the error at the previous step. For example, the present step is you are computing  $x_{k+1}$ ; how is the error at  $k+1$  step is connected to error at the  $k$  step? That is how  $\epsilon_{k+1}$  is connected to  $\epsilon_k$ . Now in the Muller method or the secant method have more than one point like  $x_k$ ,  $x_{k+1}$ , we shall see that using this particular definition we can define the  $\epsilon_{k-1}$  automatically back in terms of  $x_k$ . So finally we arrive at an expression only connecting the errors at two steps.

Now we define the order of an iterative method. The order of an iterative method is, the iterative method is of order  $p$  if error at the current step in magnitude is bounded by some constant  $C$   $\epsilon_k$  to the power of  $p$ . This  $p$  is called the order of the iterative method. It is also called the rate of convergence of the method. We will be showing that for Newton Raphson method, that  $p$  terms have to be two which means we will call the rate of convergence as two or order of convergence is two. Therefore  $\epsilon_{k+1}$  is less than equal to  $C \epsilon_k^2$ .

Similarly for Secant method we will show that this  $p$  is 1.612. Therefore we call it as super linear convergence. Now this  $C$  that is multiplying is called the asymptotic error constant. Of course we expect  $C$  is bounded quantity. When you write down the Taylor expansion you will get the value of  $C$  also. Depending on the value of  $C$  we can make few comments as to when the method may possibly fail; if  $C$  also grows and it also goes to infinity then, even though you are multiplying by a very small quantity or by a very large quantity the method can still diverge or it may converge very slowly.

(Refer Slide Time: 00:36:47 min)



The simplest of all is the Newton Raphson method. So let us take the Newton Raphson method. So the current error is  $\epsilon_{k+1}$  is  $x_{k+1}$  minus  $\xi$ . The previous error is  $\epsilon_k$  is  $x_k$  minus  $\xi$ . Now Newton Raphson method is  $x_{k+1}$  is equal to  $x_k$  minus  $f(x_k)$  by  $f'(x_k)$ . So this is the Newton Raphson method. Therefore I am substituting for  $x_{k+1}$  here i.e.  $\epsilon_{k+1}$  plus  $\xi$ ;  $x_k$  is  $\epsilon_k$  plus  $\xi$  minus  $f(x_k)$ ;  $x_k$  is again  $\epsilon_k$  plus  $\xi$   $f'(\epsilon_k + \xi)$ . Now we Taylor expand all these quantities or functions that are there. The denominator shall be taken to the numerator and then expanded. So let us open this up.

Now here you can see  $x_i$  and  $x_i$  cancels over here and so that  $\epsilon_k + 1$  is equal to  $\epsilon_k - 1$ , (when I open it and Taylor expand it) so  $f$  of  $x_i$  plus  $\epsilon_k f'$  of  $x_i$  plus half  $\epsilon_k^2 f''$  of  $x_i$  plus; so on the denominator will be  $f'$  of  $x_i$  plus  $\epsilon_k f''$  of  $x_i$ . We started with that,  $x_i$  is the exact root of our function  $f(x)$ , therefore  $f$  of  $x_i$  is zero. We shall see later on that we are talking of here is only a simple root not a multiple root. If it is a multiple root then we know that  $f'$  and  $f''$  can also be zero. So the other terms can also be zero. So let us assume for the moment we are not talking of a multiple root, we are talking of only a simple root so that  $f$  of  $x_i$  is zero,  $f'$  of  $x_i$  is not equal to zero. So I can set  $f$  of  $x_i$  is equal to zero over here.

(Refer Slide Time: 00:38:37 min)

The image shows a whiteboard with handwritten mathematical derivations. At the top, the Newton-Raphson formula is written as  $\epsilon_{k+1} + \frac{\epsilon}{2} = \epsilon_k + \frac{\epsilon}{2} - \frac{f(\epsilon_k + \frac{\epsilon}{2})}{f'(\epsilon_k + \frac{\epsilon}{2})}$ . Below this, it is simplified to  $\text{or } \epsilon_{k+1} = \epsilon_k - \frac{[f(\frac{\epsilon}{2}) + \epsilon_k f'(\frac{\epsilon}{2}) + (1/2)\epsilon_k^2 f''(\frac{\epsilon}{2}) + \dots]}{[f'(\frac{\epsilon}{2}) + \epsilon_k f''(\frac{\epsilon}{2}) + \dots]}$ . A note states "Now  $f(\frac{\epsilon}{2}) = 0$ . Therefore,". The final part of the derivation shows the simplification of the formula by canceling  $\epsilon_k$  in the numerator and denominator, resulting in  $\epsilon_{k+1} = \epsilon_k - \left[ \epsilon_k + \frac{1}{2} \epsilon_k^2 \left( \frac{f''(\frac{\epsilon}{2})}{f'(\frac{\epsilon}{2})} \right) + \dots \right] \left[ 1 + \epsilon_k \frac{f''(\frac{\epsilon}{2})}{f'(\frac{\epsilon}{2})} + \dots \right]^{-1}$ , which is then further simplified to  $\epsilon_{k+1} = \epsilon_k - \left[ \epsilon_k + \frac{1}{2} \epsilon_k^2 \left( \frac{f''(\frac{\epsilon}{2})}{f'(\frac{\epsilon}{2})} \right) + \dots \right]$ .

Now I said  $f$  of  $x_i$  is equal to zero here. If I want to take this to the numerator and expand the binomial form of one plus some capital  $x$  to the power of minus one, I will always take it as one plus  $f$  to the power of minus one or one minus  $f$  to the power of minus one form. Therefore I would take this particular quantity i.e. the first term always out. There is a constant  $f'$  of  $x_i$ , so I would take  $f'$  of  $x_i$  constant over here and you can see this term here, by taking  $f'$  as common, I will have here one plus  $\epsilon_k f''$  of  $x_i$  by  $f'$  of  $x_i$  to the power of minus one. So this quantity that I have here, I have taken it to the numerator and written as  $f'$  of  $x_i$ , the numerator  $f$  of  $x_i$  is zero. So I have written the two remaining terms  $\epsilon_k f'$  of  $x_i$  plus one by two  $\epsilon_k^2 f''$  of  $x_i$ .

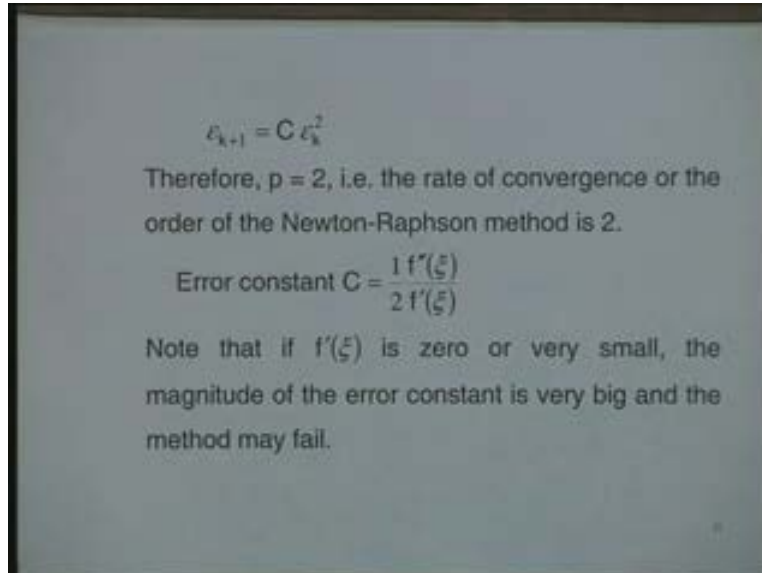
(Refer Slide Time: 00:39:29 min)

$$\begin{aligned}
 \epsilon_{k+1} &= \epsilon_k - \frac{1}{f'(\xi)} \left[ \epsilon_k f'(\xi) + \frac{1}{2} \epsilon_k^2 f''(\xi) + \dots \right] \left[ 1 + \epsilon_k \frac{f''(\xi)}{f'(\xi)} + \dots \right]^{-1} \\
 &= \epsilon_k - \left[ \epsilon_k + \frac{1}{2} \epsilon_k^2 \left( \frac{f''(\xi)}{f'(\xi)} \right) + \dots \right] \left[ 1 - \epsilon_k \frac{f''(\xi)}{f'(\xi)} + \dots \right] \\
 &= \epsilon_k - \left[ \epsilon_k + \frac{1}{2} \epsilon_k^2 \left( \frac{f''(\xi)}{f'(\xi)} \right) + \dots \right] \\
 &= \left[ \frac{1}{2} \frac{f''(\xi)}{f'(\xi)} \right] \epsilon_k^2 + O(\epsilon_k^3) = C \epsilon_k^2 + O(\epsilon_k^3)
 \end{aligned}$$

If we neglect higher order terms, we get

Therefore as I said the concept is to take the quantity that we have in the denominator to the numerator, put it to the power of minus one. Now let us assume this whole quantity as some capital X. It is easier to write it. So it is something like one plus capital X to the power of minus one. If I open it up, I will get one minus x plus x square and so on. So here I am not writing more terms because we do not need it. So it will be simply one minus capital X which is one minus epsilon k f double dash f dash plus so on. We do not require them therefore I have not written them. We have taken f dash xi inside, absorbed it into this bracket, so that f dash xi cancels with f dash xi and the second term is half epsilon k square f double dash xi by f double dash xi. So we have simplified by taking f dash xi inside in this. Now I just have to multiply these two. Now if I multiply you can see that epsilon k is epsilon k; for epsilon k square there are two terms, one into this is epsilon square, and epsilon k into epsilon k is epsilon k square. So therefore this will contribute; this will contribute; and this will contribute plus half epsilon square; this will contribute minus epsilon k square. So if I simplify I would get minus half epsilon square, whole square; f double dash minus xi plus one. Now once you multiply the whole thing I cannot simplify the whole right hand side; epsilon k cancels and I am left out with half f double dash xi by f dash xi epsilon k square. Now the other terms, I will write it in order of epsilon k cubed or I will call this term as C epsilon k square plus order of epsilon k cubed. Therefore I am able to derive that the error at the current step which xk plus one step is equal to C into epsilon k square plus order of epsilon k cube. So I now neglect this higher order terms.

(Refer Slide Time: 00:41:51)



Now we are justified in neglecting because we are saying epsilon k is a very small quantity; so order of epsilon k cubed is much smaller than order of epsilon k square. Therefore if you neglect the higher order terms, what I get here is epsilon k plus one is equal to C epsilon k square, where this constant C is half f double dash by f dash xi. Therefore we have proved that p is equal to two and that is the rate of convergence or the order of the Newton Raphson method is two.

Now when we say the Newton Raphson method is of order two what we mean is; if we are sufficiently close to the root, let us say 0.5, it means if C is sufficiently small and that C is multiplying with every epsilon; for example it is multiplying with epsilon zero square, it is multiplying with epsilon one square epsilon one square. So C is bounded quantity and it is a finite quantity. Now at each stage you are 0.5 away from the exact solution, in the next stage you will be away from 0.5 whole square i.e. 0.25. We are multiplying each one by C, the same factor. In the next step we are away from 0.25 whole square, so 0.0625.

So now see how fast the convergence will come. So you will have the accuracy coming so fast, therefore this is reasonably a good method for any practical application. So this is how initially it will take few iterations to close up. Suppose as I said the root was 1.5 and you started at 3. There is no meaning between the errors, which here is 1.5. It is quite far away. So it will take few iterations to reduce the error sufficiently small, only then you will be able to see the convergence is very fast because you have already crossed the stage wherein the epsilon k is sufficiently small. Therefore the accuracy is achieved much faster.

As I said earlier if I look at this error constant I may be able to say when the iteration may become little bit slower or when the errors initially could be big. If you can see that f dash xi is close to zero which means dy by dx slope is almost equal to zero; that means the graph is cutting the x axis almost horizontally. In that case f dash xi is going to be very small. Therefore in that case the error constant will be small but still bounded but that is going to multiply each time with this factor.

(Refer Slide Time: 00:45:33 min)

Secant method:

$$x_{k+1} = x_k - \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} f(x_k)$$

$$\epsilon_{k+1} + \xi = \epsilon_k + \xi - \frac{(\epsilon_k - \epsilon_{k-1})}{f(\xi + \epsilon_k) - f(\xi + \epsilon_{k-1})} f(\xi + \epsilon_k)$$

or  $\epsilon_{k+1} = \epsilon_k - \frac{(\epsilon_k - \epsilon_{k-1}) [ \epsilon_k f'(\xi) + (1/2) \epsilon_k^2 f''(\xi) + \dots ]}{[ \epsilon_k f'(\xi) + (1/2) \epsilon_k^2 f''(\xi) + \dots ] - [ \epsilon_{k-1} f'(\xi) + (1/2) \epsilon_{k-1}^2 f''(\xi) + \dots ]}$

Handwritten notes on the right:

$$x_{k-1} - \xi = \epsilon_{k-1}$$

$$x_k - \xi = \epsilon_k$$

$$x_{k+1} - \xi = \epsilon_{k+1}$$

Now the secant method that we have written is  $x_k$  plus one. There are two forms of secant method that we have written. One is for computation purposes and the other is for the error analysis. We shall use it -  $x_k$  plus one is equal to  $x_k$  minus  $x_k$  minus  $x_k$  minus one  $f$  of  $x_k$  minus  $f$  of  $x_k$  minus one  $f_k$ . This is the method that we have written so that error analysis can be made. The procedure is identically same. We substitute the  $x_k$  by  $\epsilon_k$  plus  $\xi$ , and then  $x_k$  minus one  $\epsilon_k$  minus one is written as in the previous step. We will write it again here;  $x_k$  minus one minus  $\xi$  is equal to  $\epsilon_k$  minus one and we have  $x_k$  minus  $\xi$  is equal to  $\epsilon_k$ . So I now substitute this and  $x_k$  plus one minus  $\xi$  is  $\epsilon_k$  plus one. Therefore I substitute  $x_k$  plus one is  $\epsilon_k$  plus one plus  $\xi$ ;  $x_k$  is  $\epsilon_k$  plus  $\xi$ ; in the numerator  $x_k$  minus  $x_k$  minus one. So I am subtracting these two, so  $\xi$  cancels off and I will be left out with only the  $\epsilon_k$  minus  $\epsilon_k$  minus one. This is  $f$  of  $\xi$  plus  $\epsilon_k$  and the second terms is minus  $f$  of  $\xi$  plus  $\epsilon_k$  minus one and this factor is  $f$  of  $\xi$  plus  $\epsilon_k$ .

Now the reason why I have chosen is,  $\xi$  here automatically cancels; one of them easily cancels here, this  $\xi$  and  $\xi$  cancels over here so that I can write this as  $\epsilon_k$  plus one  $\epsilon_k$ . I have retained the numerator as it is. Now I open up all these three terms by Taylor series;  $f$  of  $\xi$  plus  $\epsilon_k$ ,  $\epsilon_k f'(\xi)$ , half  $\epsilon_k^2 f''(\xi)$  plus one, in the denominator I have already used  $f$  of  $\xi$  is zero so we need not write  $f$  of  $\xi$  again. So we have taken  $f$  of  $\xi$  zero. The next term is  $\epsilon_k f'(\xi)$  half  $\epsilon_k^2 f''(\xi)$  plus one. Similarly  $f$  of  $\xi$  is zero  $\epsilon_k$  minus one, first derivative at  $\xi$ , half  $\epsilon_k$  minus one square  $f''(\xi)$  plus one. I have to first simplify the denominator. Write it as whatever leading factor I take out and write it as some one plus capital  $X$ . So I would like to bring it in that particular form.

(Refer Slide Time: 00:47:19 min)

$$\begin{aligned}
 \epsilon_{k+1} &= \epsilon_k - \frac{(\epsilon_k - \epsilon_{k-1}) \left[ \epsilon_k f'(\xi) + (1/2) \epsilon_k^2 f''(\xi) + \dots \right]}{(\epsilon_k - \epsilon_{k-1}) f'(\xi) + (1/2) (\epsilon_k^2 - \epsilon_{k-1}^2) f''(\xi) + \dots} \\
 &= \epsilon_k - \left[ \epsilon_k + \frac{1}{2} \epsilon_k^2 \frac{f''(\xi)}{f'(\xi)} + \dots \right] \left[ 1 + \frac{1}{2} (\epsilon_k + \epsilon_{k-1}) \frac{f''(\xi)}{f'(\xi)} \right]^{-1} \\
 &= \epsilon_k - \left[ \epsilon_k + \frac{1}{2} \epsilon_k^2 \frac{f''(\xi)}{f'(\xi)} + \dots \right] \left[ 1 - \frac{1}{2} (\epsilon_k + \epsilon_{k-1}) \frac{f''(\xi)}{f'(\xi)} \right] \\
 &= -\frac{1}{2} \frac{f''(\xi)}{f'(\xi)} [\epsilon_k^2 - (\epsilon_k^2 + \epsilon_k \epsilon_{k-1})] + \dots = C \epsilon_k \epsilon_{k-1} \\
 \text{where } C &= \frac{f''(\xi)}{2f'(\xi)}
 \end{aligned}$$

So as you can see the denominator is  $f'(\xi)$ . So I can take  $\epsilon_k \epsilon_{k-1}$  as a common factor here. Second term is half  $\epsilon_k$  square minus  $\epsilon_k \epsilon_{k-1}$ . So these two are subtracted and this is  $f''(\xi)$ . Now as I said in order to take it up I will take this particular factor out. If I take this factor out,  $\epsilon_k \epsilon_{k-1}$  is common; so it cancels off. If I take this factor out,  $\epsilon_k \epsilon_{k-1}$  cancels off. Now let us first look at the denominator. The denominator will be one plus half. This is  $\epsilon_k$  square minus  $\epsilon_k \epsilon_{k-1}$  square, factorize it, it becomes  $\epsilon_k$  minus  $\epsilon_{k-1}$  into  $\epsilon_k$  plus  $\epsilon_{k-1}$ .  $\epsilon_k$  minus  $\epsilon_{k-1}$  cancels because we are dividing by this. So what is left out is  $\epsilon_k$  plus  $\epsilon_{k-1}$ . This is  $f''(\xi)$  minus  $f'(\xi)$ , this to the power of minus one. In the numerator  $\epsilon_k$  minus  $\epsilon_{k-1}$  was cancelled. We take  $f'(\xi)$  inside. So I will have here  $\epsilon_k$  half  $\epsilon_k$  square  $f''(\xi)$  divide by  $f'(\xi)$ . Then as we have done in the previous case we will expand it by binomial expansion. This is one minus capital X. I should have written plus this this; this is one minus  $\epsilon_k \epsilon_{k-1} f''(\xi)$  by this. Now multiply it out. If I multiply it out, I would get here  $\epsilon_k$  into  $\epsilon_k$  and  $\epsilon_k$  cancels off and I will take the common factor of  $f''(\xi)$  by  $f'(\xi)$ . If I take that I will have minus sign outside,  $\epsilon_k$  square here and I will have here  $\epsilon_k$  into this factor which means  $\epsilon_k$  into this is  $\epsilon_k$  square  $\epsilon_k$  into  $\epsilon_{k-1}$ . So the product of these two is this and this. Now if I simplify this I would get here  $C \epsilon_k \epsilon_{k-1}$ , where this constant is  $f''(\xi)$  by twice  $f'(\xi)$ . This constant is same as in the Newton Raphson method but it has got a different form.

We will see now how we are going to manipulate this from this to get the order of this particular method; because when we have defined the order it is to link only  $\epsilon_{k+1}$  and  $\epsilon_k$  but here it has linked  $\epsilon_{k+1}$  with  $\epsilon_k$  and  $\epsilon_{k-1}$ . So we will have to eliminate  $\epsilon_{k-1}$  from here to get that particular step and we have the solution.